



Parallel MuSES for Infrared Signature Modeling of U.S. Army Vehicles and Targets

by Rama R. Valisetty, Raju R. Namburu, and George Petit

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Parallel MuSES for Infrared Signature Modeling of U.S. Army Vehicles and Targets

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14. ABSTRACT The objective for this work is the parallelization of MuSES, a commercially available computer code for infrared signature modeling of U.S. Army vehicles and targets. During an earlier phase of development in 2002, the progress in this work included (1) profiling of the code for identifying potential segments of thermal and radiation computations for parallelization, (2) selecting a normalization loop within the selective over relaxation (SOR) iteration loop for parallelization by OpenMP, and (3) some results for the ZSU Quad 23 AA vehicle. The results then indicated a scalability performance of some 55% reduction in wall clock computation time on eight processors. Since then, the following parallelization enhancements were added to the code: (1) partial direct solution computation loop within the SOR iteration loop, (2) tri-diagonal solution computation loop within the SOR iteration loop, (3) accelerated solution computation loop within the SOR iteration loop, and (4) routine ComputeRadNodeAveValues. The effectiveness of the enhanced code was evaluated on the previously mentioned ZSU vehicle, a car, and the M2 Bradley Fighting Vehicle. As a result of the enhancements, the maximum performance improvement of the code increased to an average 72% reduction in the wall clock time on 16 processors for computing radiance in the IR range (3–4 μm , 8–12 μm).					
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Contents

List of Figures	iv
List of Tables	iv
1. Background	1
2. Parallelization Approach	1
3. Code Segments for Parallelization	2
4. Parallelization Details	3
4.1 Load Balancing.....	5
4.2 Thermal Node Temperature Change Summation and Maximums.....	5
4.3 Recursions Within Routines Defined in Parallel Regions.....	5
4.4 Maintaining Thread-Safe Code	5
5. Verification of the Results	6
6. Scalability of the Results	6
7. Future Parallelization Work	8
Distribution List	9

List of Figures

Figure 1. Serial MuSES solver profile.....	3
Figure 2. Flow diagram for Nodal Network solver.....	4
Figure 3. Physical temperature computed at 1900 by the original serial code and the parallel code.....	6
Figure 4. Prius car mesh.....	7
Figure 5. M2 BFV.....	7

List of Tables

Table 1. Mesh details.	7
Table 2. Wall clock run times with the initial phase of parallelization.	8
Table 3. Wall clock run times with the second phase of parallelization.....	8

1. Background

MuSES is an acronym for the Multi-Service Electro-Optical Signature Code. While this code is relatively new in Department of Defense (DOD) communities, its predecessors, most notably, the Physically Reasonable Infrared Signature Model on which it is based, have a long and successful history of use in the thermal infrared (IR) modeling of ground vehicles and targets. The code grew mainly out of the need for integrating the many legacy codes and also to make use of the new computing and software programming techniques such as configuration management, etc.

While there are many sponsors from the DOD service agencies from the U.S. Army side, it is the U.S. Army Tank-automotive and Armaments Command Tank-Automotive Research, Development, and Engineering Center which has sponsored the development of the code for a number of years and is currently conducting the validation and verification of version 6.0 of the code. A commercial version of the code, RADTHERM, is widely used (sans radiation) in automotive industries. Both MuSES and RADTHERM are being developed commercially by ThermoAnalytics of Calumet, MI. Help in building and running MuSES models is available.¹⁻³

2. Parallelization Approach

For the present work, a version of the source code was obtained from ThermoAnalytics, Inc. The present work was devoted to making the MuSES code run on parallel computers. This was achieved by using OpenMP.* OpenMP is a shared-memory programming standard with parallel regions of computation explicitly identified. This calls for implicit messaging and explicit synchronization. It usually runs with one master process and many slave computing processes and is best suited to run on shared-memory architectures such as SGI Origin 3000* computers. Profiling the source code on an application of interest usually identifies subroutines and code segments that are good candidates for parallelization with OpenMP threads. Only a limited amount of computational scalability is generally achieved with this approach.

The other approach that can lead to a better scalability is based on the message-passing interface (MPI) standard. MPI is usually implemented with domain decomposition. Scalability is

¹Makens, C.; Wood, B. *MuSES 6.0 User's Manual*; ThermoAnalytics, Inc.: Calumet, MI, 2001.

²*MuSES 6.0 Training Manual*; ThermoAnalytics, Inc.: Calumet, MI, 2000.

³Gonda, T.; Polsen, E.; Jones, J.; Sass, D.; White, J.; Gerhardt, A.; Less, D.; Rynes, P. *A Report on the V&V of MuSES 6.0*; draft report; U.S. Army Tank-Automotive Research, Development, and Engineering Center: Warren, MI.

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achieved in numerical codes containing computational meshes, by partitioning the meshes into several submeshes or computational domains. Each domain is assigned to a different processor. Elements internal to a domain are computed independently of calculations on other processors; elements on the interdomain boundaries are computed with neighboring processors exchanging the relevant information in explicit messages. Synchronization of computations (i.e., for time steps, etc.) is a code writer's responsibility. Since MPI doesn't require shared-memory architecture, code segments have to be selected with much more understanding of the computational algorithms in a source code rather than with simple profiler knowledge alone.

OpenMP was thus selected for the present work because it allowed the authors to quickly start the parallelization process. Using MPI would have meant more studies of the MuSES computational modules. The code's many physical processes are coupled and nonlinear. The authors wanted to quickly show that a parallelization of MuSES is possible and a thorough application of either OpenMP or MPI would extend the thermal IR modeling problem space by allowing DOD to leverage its high-performance computing resources for increasing the throughput and lowering the wall clock time.

Results are provided to show the efficiency achieved by the parallel code. Results for physical and apparent temperatures and radiance were computed for the ZSU Quad 23 AA vehicle under four different paint conditions and were verified for accuracy first. For establishing the scalability of the multithreaded code, results were generated from 2-, 4-, 8- and 16- multiprocessor runs. Parallelization work proceeded in two phases. The initial phase focused on a "normalization loop within the selective over relaxation (SOR) iteration loop." With this parallelization, reduction in the range of 30%–40% in wall clock time in the solver routines was achieved for up to four processors. These results were reported by Petit et al.⁴ Parallelization of three additional loops in the "SOR iteration loop" and some work in the "ComputeRadNodeAveValues" routine improved the scalability to ~88% reduction in computational time with 16 processors using a test model case.

3. Code Segments for Parallelization

Parallelization by OpenMP multithreads is a gradual process. First, candidate subroutines and code segments are to be determined for opportunities for parallelization from a profile of times spent in various subroutines, as shown in figure 1. The "Nodal Network" solver was identified as a potential candidate subroutine. This subroutine computes for 72.5% of the total time. The parallelization effort, therefore, naturally focused in this area. The flowchart for this routine

⁴Petit, G.; Valisetty, R.; Namburu, R. Parallel MuSES for Infrared Signature Modeling of Army Vehicles and Targets. Presented at the 13th Annual Ground Target and Validation (GTM&V) Conference, Houghton, MI, August 2002.

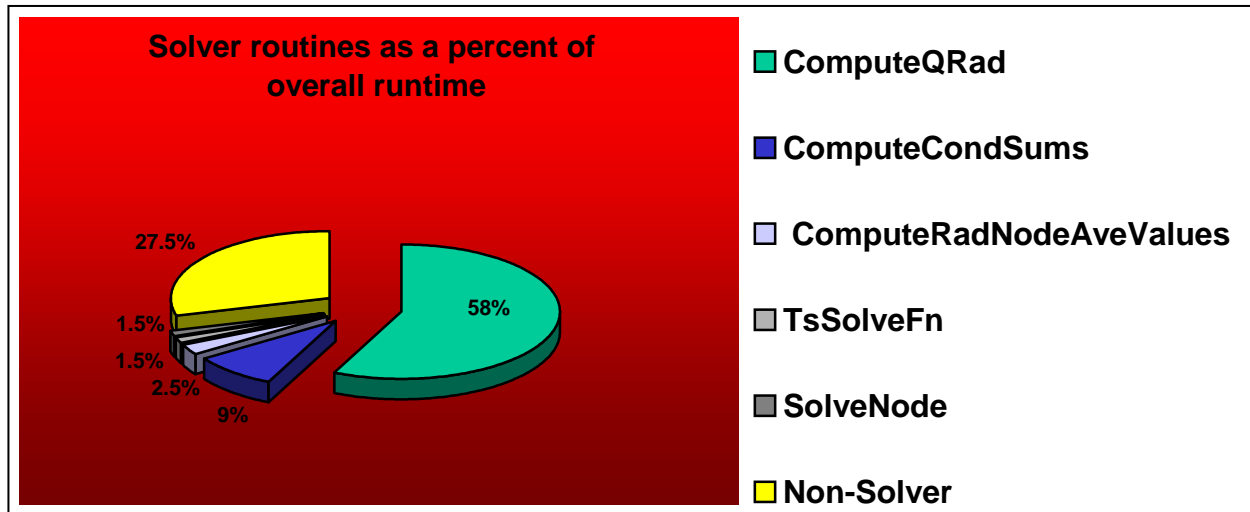


Figure 1. Serial MuSES solver profile.

is provided in figure 2. In this figure, the computational regions with parallel threads are indicated with a “p” in the upper right-hand corners.

4. Parallelization Details

Based on the code profiling performed, attention was focused on the Nodal Network solver. It begins with initialization and sets up two main loops for time stepping and SOR convergence iteration within each time step. Within each SOR convergence iteration loop, the following major computational steps are performed before evaluating the convergence criteria:

1. Average radiation nodes computation.
2. Partial direct solution computation loop for thermal nodes.
3. Tri-diagonal solution computation loop for thermal nodes.
4. Accelerated solution computation loop for thermal nodes.
5. Normalized solution computation loop for thermal nodes.

Steps 2–5 eventually utilize routines ComputeCondSum and ComputeQRad, which comprise the bulk of the computational time used in the solver. Both of these routines are highly scalar and serial in nature, making them poor candidates for parallelization. Hence, the higher-level loop structures which eventually call these routines became the focus of the parallelization effort. The following subsections summarize the parallelization issues that were encountered and include a synopsis of the implemented solutions.

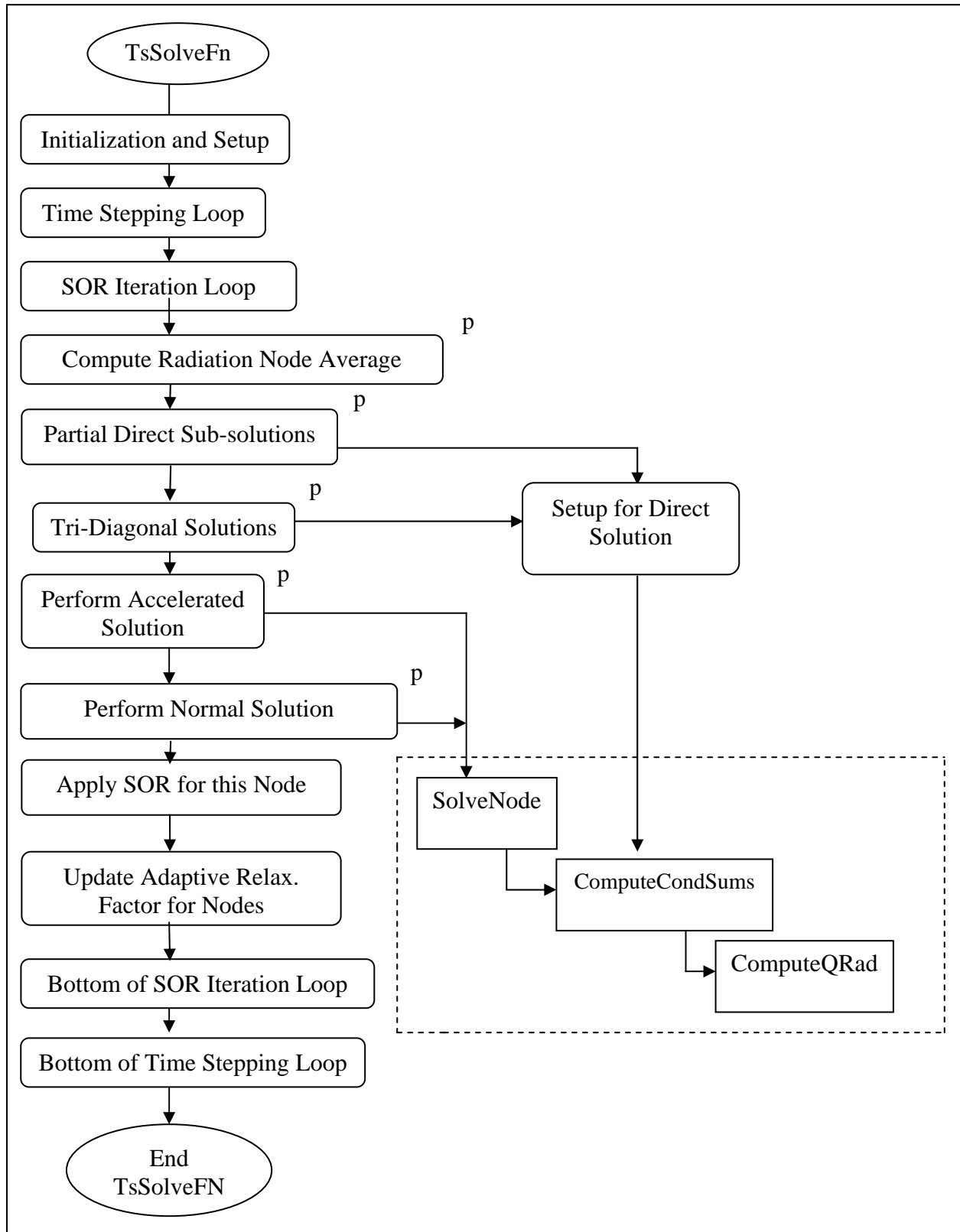


Figure 2. Flow diagram for Nodal Network solver.

4.1 Load Balancing

For the Partial-Direct, Tri-Diagonal and Accelerated solution loops, only a subset of the thermal nodes are processed. For the first two cases, the original code would loop through all thermal nodes and check a variable in the thermal-node structure to determine if that node was defined as appropriate for further processing. This logic could introduce significant load-balancing problems when distributing the calculations across all the thermal nodes within the loop. This problem was overcome by inserting logic in the setup portion of the solver that created separate arrays containing a list of only those thermal nodes being used for partial-direct or tri-diagonal solutions, respectively. The loop structure within the SOR convergence loop was then modified to loop over these new arrays instead of the entire thermal node array. However, some load imbalance is still inherent in the partial-direct loop because each partial-direct thermal node can be associated with 2–24 other nonpartial-direct thermal nodes. This leads to a varying amount of work to migrate the new thermal node values from the partial-direct thermal node to the nonpartial-direct thermal nodes associated with it. For the accelerated computation case, this same solution had already been implemented within the code.

4.2 Thermal Node Temperature Change Summation and Maximums

A product of the Accelerated and Normal solution computations are summations and maximums of all node temperatures, temperature changes, and residuals. The summations were implemented using an OpenMP Reduction clause. Maximums of the temperature changes and residuals were computed for each processor in a parallel fashion in the “SolveNode” subroutine of the block and saved in separate elements of a shared array. The overall maximum was determined by the master process after the parallel computation completed.

4.3 Recursions Within Routines Defined in Parallel Regions

In the “ComputeCondSums” subroutine, nonnative thermal-node temperatures were accessed on the read-only basis so as to preserve the recursive computations that were taking place in it. Also, the code provides the capability to associate multiple thermal nodes with a single radiation node. Because of this capability, ComputeQRad and ComputeRadNodeAveValues update radiation nodes associated with multiple thermal nodes in OpenMP critical regions. This requirement does have performance implications for models utilizing significant numbers of multiply-associated radiation nodes. It is, therefore, recommended that the use of this feature be avoided as much as possible when developing models that will be used by parallel versions of the MuSES code.

4.4 Maintaining Thread-Safe Code

Two nonsolver routines, fluidtable and xydata, modified global (shared) variables that led to erroneous results being generated. These routines were modified to use private copies of shared variables.

5. Verification of the Results

For generating the results to verify the accuracy of the included OpenMP multithreads, the ZSU Quad 23 AA vehicle was considered. It was assumed to be at 47.175° latitude and 88.492° longitude, and at a 183.49-m elevation, on 19 July 1984, exposed to ambient atmospheric conditions for 6 hr beginning at 1300. It was considered with four different paint conditions (desert tan, rust, black, and tactical glass).

A mesh for the vehicle was available with 2339 elements, 8853 vertices, and 128 parts; it was selected. Results for the physical and apparent temperatures and radiance ($3\text{--}4\text{ }\mu\text{m}$, $8\text{--}12\text{ }\mu\text{m}$) were computed with both the original serial code and the code with the OpenMP multithreads in a 4-processor run. Results from the two codes were identical. Physical temperature results at 1900 are shown in figure 3. Results for the apparent temperature and the $8\text{--}12\text{ }\mu\text{m}$ results were also similarly identical, thus verifying that the OpenMP multithreads were properly inserted.

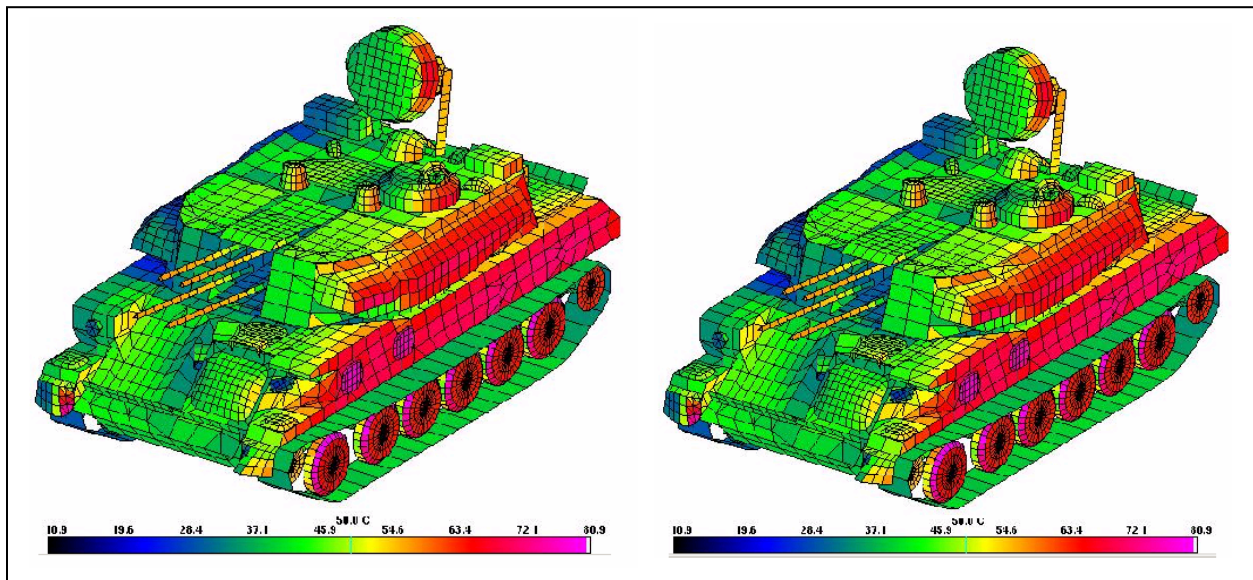


Figure 3. Physical temperature computed at 1900 by the original serial code and the parallel code.

6. Scalability of the Results

Scalability results are generated on three meshes. The first mesh is the previously mentioned ZSU Quad 23 AA vehicle. The other two are a car and the M2 Bradley Fighting Vehicle (BFV) as shown in figures 4 and 5. The details of the three meshes are listed in table 1.

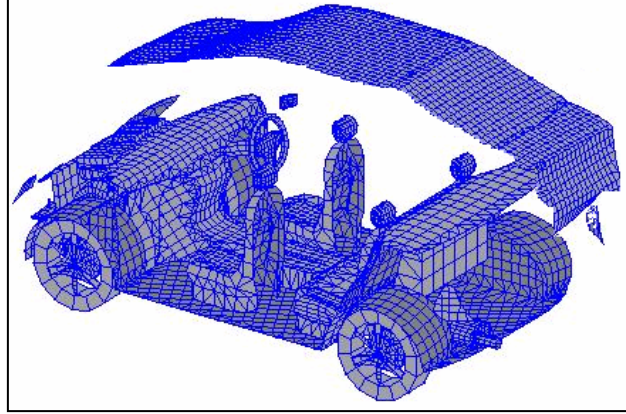


Figure 4. Prius car mesh.

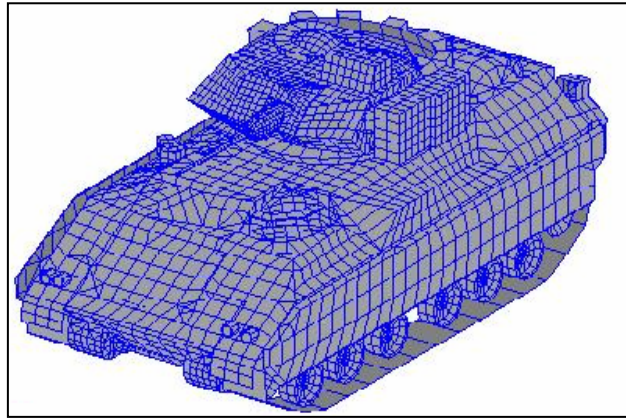


Figure 5. M2 BFV.

Table 1. Mesh details.

Vehicle	Elements	Vertices	Parts	Thermal Nodes
Prius car	18707	16234	33	29709
M2 BFV	9411	9048	100	22058
ZSU tank	2339	8853	128	18351

Simulation times for the Prius, M2, and ZSU meshes were 1800, 36,000, and 25,200 s, respectively. All were subjected to engine and other thermal fluxes and ambient conditions. All three meshes were run with 1, 4, 8, and 16 processors on an SGI Origin 3000 machine in a shared-memory environment. Wall clock times which are significant for users are presented in two sets: (1) results in table 2 with the initial parallelization and (2) results in table 3 with the second set of parallelization enhancements.

Table 2. Wall clock run times with the initial phase of parallelization.

Vehicle	1 proc run (s)	4 proc run (s)	8 proc run (s)	16 proc run (s)
Prius car	16351	8335	7581	7702
ZSU tank	187598	123179	107729	110436
M2 BFV	256898	217163	239927	147084

Source: footnoted reference 4.

Table 3. Wall clock run times with the second phase of parallelization.

Vehicle	1 proc run (s)	4 proc run (s)	8 proc run (s)	16 proc run (s)
Prius car	16351	4746	3825	3334
ZSU tank	187598	63458	62665	64545
M2 BFV	256898	95176	90516	76476

With the initial phase of parallelization, the reductions were significant for the 1–4 processor runs. However, reductions were not maintained from the 8–16 processor runs indicating that scalability was not maintained. This is to be expected, however, since only a part of the MuSES code was parallelized.

But with the second phase parallelization of the four loops, the scalability improved to an average 72% reduction in computation time with 16 processors, as can be seen from table 3.

Scalability of results is also affected by how the runs were made. Since the runs were made in a shared-memory environment, wall clock times are usually affected by other jobs running at the run time. For a strict comparison of run times, jobs must be run in a dedicated bench mark queue. This was not done, however, since such queues are not available to typical users.

7. Future Parallelization Work

Future efforts in this project will include porting the OpenMP parallel version of MuSES to the IBM SP4 platform and converting the parallel version to a message-passing (i.e., MPI) paradigm in order to provide more scalability and portability.

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